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**A STATISTICAL TECHNIQUE
FOR COMPUTER IDENTIFICATION
OF OUTLIERS IN MULTIVARIATE DATA**

by

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16. Abstract A statistical technique and the necessary computer program for editing multivariate data are presented. The technique is particularly useful when large quantities of data are collected and the editing must be performed by automatic means. One task in the editing process is the identification of outliers, or observations which deviate markedly from the rest of the sample. A statistical technique, and the related computer program, for identifying the outliers in univariate data was presented in NASA TN D-5275. The current report is a multivariate analog which considers the statistical linear relationship between the variables in identifying the outliers. The program requires as inputs the number of variables, the data set, and the level of significance at which outliers are to be identified. It is assumed that the data are from a multivariate normal population and the sample size is at least two greater than the number of variables.			
Although the technique has been used primarily in editing biodata, the method is applicable to any multivariate data encountered in engineering and the physical sciences.			
An example is presented to illustrate the technique.			
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A STATISTICAL TECHNIQUE FOR COMPUTER IDENTIFICATION OF OUTLIERS IN MULTIVARIATE DATA

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INTRODUCTION

The NASA Flight Research Center is engaged in an extensive biomedical research and development program. Objectives of this program include advancing the state of the art in the medical monitoring of humans in flight (ref. 1); predicting and extending the limit of man's operational capacity in the flight environment; and developing improved protection, restraint, and life support systems. As a result of this program, large quantities of biomedical information are collected in flight, necessitating dependence on the Flight Research Center's capacity for collecting, reducing, and analyzing these data by automatic means.

Experience has shown that no matter how sophisticated the monitoring, collection, and reduction systems, some editing of the biodata is required before they can be analyzed statistically. The reduced biodata may contain observations that deviate markedly from the rest of the sample. Such observations may be due to errors other than the usual random fluctuations characterizing the population to which the data belong, or may merely occur too infrequently to be considered in a particular analysis. If, upon examination, an observation falls outside a standardized region, it is usually identified as an outlier. Outliers often provide useful information. Their identification is important not only for improving the analysis but also for indicating anomalies which may require further investigation.

A statistical technique, and the related computer program, for identifying the outliers in univariate data was presented in reference 2. A method for identifying outliers in multivariate data is derived and demonstrated in this report. This method was chosen because of its simplicity and applicability in editing biodata. A program for automatic editing was written in FORTRAN IV. Inputs to this program are the number of variables, the data set, and the selected level of significance. An example is presented to illustrate the use of the method, and a scatter plot of the data is shown. The program source listing, user instructions, and a sample output are also presented.

The program computes and prints the means and standard deviations of all the variables before and after the outliers are identified and deleted. A list of the data with outliers identified by asterisks is also printed.

The authors would like to acknowledge the assistance of M. C. Nesel in writing the computer program.

SYMBOLS

A	nonsingular matrix
$F_{\alpha;p, n-p-1}$	α -level value of F-distribution with p and $(n - p - 1)$ degrees of freedom
G	normal component of acceleration as experienced by the subject, g
H/R	heart rate, beats per minute
I	identity matrix
\sum_i^k	summation starting from i through k, where i and k are integers between 1 and n, and i is less than k
$N_p(\mu, \Gamma)$	p-variate normal distribution with mean, μ , and covariance matrix, Γ
n	sample size
p	number of variables
S	$(p \times p)$ matrix of sums and cross-products of deviations of observations from X divided by $(n - 1)$
S.D.	standard deviation
T^2	Hotelling's T^2 statistic
u, v	column vectors of p dimensions
X_i, X_j	i th or j th observation vector of p dimensions, where i or j ranges from 1 to n
\bar{X}	mean vector computed from n observation vectors
Z_i	i th vector obtained by orthogonal transformation of vector X_i
α	level of significance
Δ_i	positive real number corresponding to observation vector X_i , where i ranges from 1 to n
Δ_*	positive real number computed from $F_{\alpha;p, n-p-1}$ for the data set, to compare with Δ_i

τ^2 random variable related to T^2

Superscript:

T

transpose

BRIEF DESCRIPTION OF TECHNIQUE

Outliers are identified by computing, at the given level of significance, the critical value, Δ_* , for the data set and Δ_i for each observation vector, X_i . If Δ_i is larger than Δ_* , observation X_i is identified as an outlier. The quantity Δ_* is a function of total sample size, n , number of variables, p , and the F-value for the given level of significance, whereas each Δ_i is a function of the observation X_i and the estimated mean and covariance matrix from all the observations. It is assumed that all the observations constitute a random sample from a p-variate normal distribution.

DERIVATION OF TECHNIQUE

Let X_1, X_2, \dots, X_n be a random sample of size n from a p-dimensional normal distribution, $N_p(\mu, \Gamma)$. The observations will be considered as n greater than $p + 2$ column vectors in a p-dimensional vector space. Consider any $(n \times n)$ orthogonal matrix, with first two rows as shown,

$$\begin{bmatrix} \frac{1}{\sqrt{n}} & \frac{1}{\sqrt{n}} & \cdots & \frac{1}{\sqrt{n}} \\ \sqrt{\frac{n-1}{n}} & -\frac{1}{\sqrt{n(n-1)}} & \cdots & -\frac{1}{\sqrt{n(n-1)}} \\ \cdot & \cdot & \cdots & \cdot \\ \cdot & \cdot & \cdots & \cdot \end{bmatrix}$$

representing the rotation of n -dimensional space so that the observations X_1, X_2, \dots, X_n are transformed into vectors Z_1, Z_2, \dots, Z_n , where

$$Z_1 = \sum_{i=1}^n \frac{1}{\sqrt{n}} X_i = \sqrt{n} \bar{X}$$

$$Z_2 = \sqrt{\frac{n-1}{n}} X_1 - \sum_{i=2}^n \frac{1}{\sqrt{n(n-1)}} X_i = \sqrt{\frac{n}{n-1}} (X_1 - \bar{X})$$

It may be noted that Z_1 is distributed as $N_p(\sqrt{n}\mu, \Gamma)$, Z_2, Z_3, \dots, Z_n are all distributed as $N_p(0, \Gamma)$, and all are stochastically independent of one another (ref. 3, pp. 50-52). Let S denote the estimate of the covariance matrix Γ . Then the following relation holds:

$$(n - 1)S = \sum_{i=1}^n (X_i - \bar{X})(X_i - \bar{X})^T = \sum_{i=2}^n Z_i Z_i^T$$

Define a $(p \times p)$ matrix, S_o , such that

$$(n - 2)S_o = \sum_{i=3}^n Z_i Z_i^T$$

Independence of Z_2 and set (Z_3, \dots, Z_n) implies that Z_2 is independent of S_o and the

$$T^2 = Z_2^T S_o^{-1} Z_2 \quad (1)$$

statistic is distributed as Hotelling's T^2 . From the relationship between T^2 and F (ref. 3, pp. 106-107), it follows that

$$T^2 = Z_2^T S_o^{-1} Z_2$$

is distributed as

$$\frac{p(n - 2)}{(n - p - 1)} F_{p, n - p - 1}$$

Because Z_2 is not independent of S , the preceding distribution does not hold for

$$\tau^2 = Z_2^T S^{-1} Z_2 \quad (2)$$

and the distribution of τ^2 must be derived.

By the preceding definitions

$$(n - 1)S = Z_2 Z_2^T + \sum_{i=3}^n Z_i Z_i^T = Z_2 Z_2^T + (n - 2)S_o$$

or

$$(n - 2)S_O = (n - 1)S - Z_2 Z_2^T \quad (3)$$

To express the relation between T^2 and τ^2 the following lemma is used:

Lemma: Let A be a $(p \times p)$ nonsingular matrix and u, v be p -dimensional vectors. Then

$$(A - uv^T)^{-1} = A^{-1} + \frac{(A^{-1}u)(v^T A^{-1})}{1 - v^T A^{-1}u} \quad (4)$$

Proof: The proof of the lemma is presented in appendix A.

Applying the result (eq. (4)) of the lemma to equation (3),

$$S_O^{-1} = \frac{n-2}{n-1} \left[S^{-1} + \frac{S^{-1}Z_2 Z_2^T S^{-1}}{(n-1) - Z_2^T S^{-1} Z_2} \right]$$

Substituting this expression for S_O^{-1} in equation (1) and applying equation (2),

$$\begin{aligned} T^2 &= Z_2^T S_O^{-1} Z_2 = \frac{n-2}{n-1} Z_2^T \left[S^{-1} + \frac{S^{-1}Z_2 Z_2^T S^{-1}}{(n-1) - Z_2^T S^{-1} Z_2} \right] Z_2 \\ &= \frac{n-2}{n-1} \left[Z_2^T S^{-1} Z_2 + \frac{1}{(n-1) - Z_2^T S^{-1} Z_2} (Z_2^T S^{-1} Z_2 Z_2^T S^{-1} Z_2) \right] \\ &= \frac{n-2}{n-1} \left[\tau^2 + \frac{\tau^4}{(n-1) - \tau^2} \right] \\ &= \frac{(n-2)\tau^2}{(n-1) - \tau^2} \end{aligned} \quad \left. \right\} (5)$$

This relation provides the distribution of τ^2 , and appropriate probability statements can be made.

Define

$$\Delta_i = (X_i - \bar{X})^T S^{-1} (X_i - \bar{X}) \text{ for } i = 1, 2, \dots, n$$

With no loss in generality, Δ_1 is used. From equation (2)

$$T^2 = Z_2^T S^{-1} Z_2 = \frac{n}{n-1} \Delta_1$$

and from equation (5)

$$\begin{aligned} T^2 &= \frac{(n-2)\frac{n}{n-1}\Delta_1}{(n-1) - \frac{n}{n-1}\Delta_1} \\ &= \frac{n(n-2)\Delta_1}{(n-1)^2 - n\Delta_1} \end{aligned}$$

is distributed as

$$\frac{p(n-2)}{(n-p-1)} F_{p, n-p-1}$$

From the distribution of T^2 , the statement

$$\text{Probability} \left[\frac{n(n-2)\Delta_1}{(n-1)^2 - n\Delta_1} \geq \frac{p(n-2)}{(n-p-1)} F_{\alpha; p, n-p-1} \right] = \alpha$$

provides criteria for identifying the Z_2 (or X_1) as an outlier at the assigned level of significance, α . This statement is equivalent to

$$\text{Probability} \left[\Delta_1 \geq \frac{p(n-1)^2 F_{\alpha; p, n-p-1}}{n(n-p-1) + np F_{\alpha; p, n-p-1}} \right] = \alpha$$

For significance level α , denote

$$\Delta_* = \frac{p(n-1)^2 F_{\alpha; p, n-p-1}}{n(n-p-1) + np F_{\alpha; p, n-p-1}}$$

then X_1 will be identified as an outlier at α level if

$$\Delta_1 > \Delta_*$$

The quantity X_1 (or Z_2) was chosen for convenience of the preceding derivation and the derivation holds for all X_i . Thus X_i will be identified as an outlier at α level of significance if $\Delta_i > \Delta_*$.

PROGRAM APPLICATION

Given a sample of data vectors X_1, X_2, \dots, X_n , the mean vector

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$

and the estimate of the covariance matrix

$$S = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})(X_i - \bar{X})^T$$

is computed. Then, for assigned α , the critical value

$$\Delta_* = \frac{p(n-1)^2 F_{\alpha; p, n-p-1}}{n(n-p-1) + np F_{\alpha; p, n-p-1}}$$

for the data set is computed. Corresponding to each observation vector, X_i ,

$$\Delta_i = (X_i - \bar{X})^T S^{-1} (X_i - \bar{X})$$

is computed. If $\Delta_i > \Delta_*$, observation vector X_i is identified as an outlier at level α .

The program (appendix B) follows this technique. The output (appendix C) of the program contains the data set, Δ_i , Δ_* , outliers marked by asterisks (*), the number of outliers identified, and the level of significance. The output also shows the means and standard deviations of the variables before and after the deletion of outliers. The required input parameters are: (1) format of the data to be read, (2) number of variables, (3) significance level, α , and (4) the data set, formatted as specified. Program options allow the user to select either a 5 percent or a 1 percent level of significance and to print the names of the variables, if desired. This program is designed so that it can be used as a subroutine in other than biodata applications, in engineering and the physical sciences, for example.

The program is particularly useful when large quantities of data are collected and the editing must be performed by automatic means.

EXAMPLE

Heart rate, H/R, and normal component of acceleration, G, data from a 66-minute flight by a student pilot at the Aerospace Research Pilot School, Edwards Air Force Base, Calif., are used to demonstrate the described technique of computer editing of biodata. These data were chosen because centrifuge studies (ref. 4) have shown that H/R and G are linearly related. The program was used to identify the outliers at a 1 percent level of significance considering H/R and G separately as univariate data and together as bivariate data. The computer output for these cases is shown in appendix C.

The results of the two univariate analyses and the one bivariate analysis of the same data are presented in figure 1. The point labeled H is identified as an outlier on the basis of H/R analysis alone; the point labeled G is identified as an outlier on the basis of G alone; and points labeled B are identified as outliers on the basis of bivariate analysis of H/R and G.

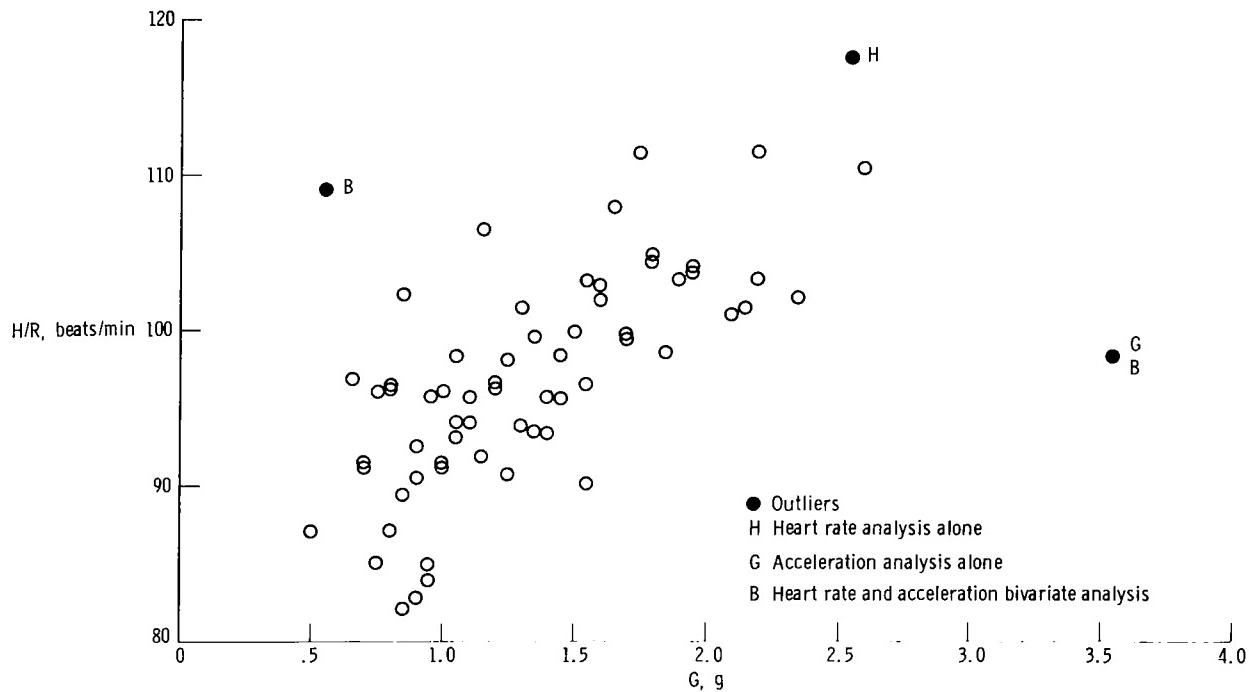


Figure 1. Minute heart rate and acceleration data for a 66-minute flight of a student pilot from the Aerospace Research Pilot School showing outliers identified by the automatic multivariate outlier technique.

Bivariate analysis is based on the fact that high H/R is associated with high G, whereas univariate analysis cannot take this information into account. For this reason the point labeled H was not identified by the bivariate analysis, but was identified as an outlier on the basis of univariate H/R analysis. Also, both points labeled B appear not to follow a statistical linear relationship and are identified by the bivariate

analysis; however, only one of these points was identified by one of the univariate analyses (G alone). This example thus focuses on the fact that the multivariate technique, which utilizes the statistical linear relationships between the variables, is preferable in identifying outliers in multivariate data.

CONCLUDING REMARKS

A statistical technique to identify outliers, or observations which deviate markedly from the rest of the sample, in multivariate data at a given level of significance was derived. The use of the technique was illustrated by a biodata example. The example also demonstrated that the results obtained when each variable was considered separately could be different from the results obtained when the variables were considered jointly. The latter technique takes into account the statistical linear relationship between the variables and is the preferred method.

Although this method of detecting and identifying outliers is being used for biodata editing at the NASA Flight Research Center, it is also applicable to multivariate data encountered in other disciplines, such as engineering and the physical sciences. This technique is particularly useful when large quantities of data are collected and the editing must be performed by automatic means.

The program can be used as a subroutine in multivariate analyses.

Flight Research Center,
National Aeronautics and Space Administration,
Edwards, Calif., May 5, 1971.

APPENDIX A

PROOF OF THE LEMMA

Lemma: If A is a $(p \times p)$ nonsingular matrix, and u, v are p -dimensional vectors, then

$$(A - uv^T)^{-1} = A^{-1} + \frac{(A^{-1}u)(v^T A^{-1})}{1 - v^T A^{-1}u}$$

Proof: The result is obtained by showing that

$$(A - uv^T) \left[A^{-1} + \frac{(A^{-1}u)(v^T A^{-1})}{1 - v^T A^{-1}u} \right] = I$$

Simplification of the left-hand expression gives

$$AA^{-1} + \frac{AA^{-1}u(v^T A^{-1})}{1 - v^T A^{-1}u} - uv^T A^{-1} - \frac{uv^T A^{-1}uv^T A^{-1}}{1 - v^T A^{-1}u}$$

or

$$I + \frac{1}{1 - v^T A^{-1}u} \left[uv^T A^{-1} - uv^T A^{-1} + (v^T A^{-1}u)(uv^T A^{-1}) - uv^T A^{-1}uv^T A^{-1} \right]$$

Because $v^T A^{-1}u$ is a scalar, the expression becomes

$$I + \frac{1}{1 - v^T A^{-1}u} \left[(v^T A^{-1}u)(uv^T A^{-1}) - (v^T A^{-1}u)(uv^T A^{-1}) \right]$$

or

$$I$$

which is the same as the right-hand side.

APPENDIX B

PROGRAM SOURCE LISTING

```
C MTVOUT
C
C PURPOSE
C     IDENTIFY OUTLIERS IN MULTIVARIATE DATA
C
C METHOD
C     LET VECTORS X(1) THRU X(N) BE OBSERVATIONS FROM A NP-VARIATE
C     NORMAL DISTRIBUTION
C     THEN
C         XBAR AND COVARIANCE MATRIX S ARE OBTAINED BY EQUATIONS
C         XBAR = (1/N) * SUM X(J)
C         S * (N-1) = SUM(X(J)-XBAR) * (X(J)-XBAR) TRANSPOSE
C
C     WHERE
C         N = SAMPLE SIZE          (N.LE.500)
C         NP = NUMBER OF VARIABLES (NP.LE.10)
C         ALPH = SIGNIFICANCE LEVEL
C         FALPH = F-VALUE FOR ALPHA AT NP AND N-NP-1
C                  DEGREES OF FREEDOM
C         X(J,I) = THE 'I' TH ELEMENT OF THE 'J' TH VECTOR,
C                  WHERE I=1,2,...,NP AND J=1,2,...,N
C
C     CALCULATE
C         DELSTAR = ((N-1)**2*NP*FALPH) / (N*((N-NP-1)+NP*FALPH))
C
C     CALCULATE FOR EACH OBSERVATION VECTOR X(J)
C         RR = (X(J)-XBAR) TRANSPOSE * (S) INVERSE * (X(J)-XBAR)
C
C     IF RR > DELSTAR, THEN X(J) IS IDENTIFIED AS AN OUTLIER
C
C REFERENCE
C
C 1. AN INTRODUCTION TO MULTIVARIATE STATISTICAL ANALYSIS,
C    ANDERSON, 1965
C 2. A SIMPLE TECHNIQUE FOR AUTOMATIC COMPUTER EDITING
C    OF BIODATA, NASA TN D-5275
C 3. SYS/360 SCIENTIFIC SUBROUTINE PACKAGE (360A-CM-03X)
C    PROGRAMMERS MANUAL, IBM INC., 1968
C
C SUBROUTINES
C
C     FTABLE
C     MPRD      (IBM SSP)
C     LOC       "   "
C     DSINV     "   "
C     DMFSD    "   "
C
```

APPENDIX B

C INPUT

```

C CARD 1 FORMAT OF X-ARRAY CARDS TO BE READ IN      (20A4)
C CARD 2
C     COL 1- 2      NP    (NUMBER OF VARIABLES)  NP.LE.10   (I2)
C     COL 3        BLANK
C     COL 4- 6      ALPH   (SIGNIFICANCE LEVEL) .05 OR .01 (F3.2)
C     COL 7- 9      BLANK
C     COL 10       VARIABLE NAME CARD INDICATOR
C                  1 - NAME CARD FOLLOWS
C                  BLANK - NO NAME CARD
C CARD 3 (OPTIONAL)
C     TEN FIELDS OF EIGHT CHARACTERS EACH (10A8), WHICH MAY
C     BE USED TO ASSIGN MEANINGFUL NAMES TO THE NP VARIABLES.
C     IF COL 10 OF THE PREVIOUS CARD IS PUNCHED, NAMES MUST
C     BE ASSIGNED FOR ALL NP VARIABLES. DEFAULT NAMES ARE
C     'X1', 'X2', ... , 'X(NP)'.
C CARDS 4-      DATA FORMATTED AS PRESCRIBED IN CARD 1
C
C MULTIPLE RUNS ARE PERMITTED, AS LONG AS EACH DATA DECK IS
C PRECEDED BY APPROPRIATE CONTROL CARDS (CARDS 1, 2 AND
C 3 ABOVE), AND IS FOLLOWED BY A CARD WITH *** PUNCHED
C IN COLUMNS 1 THRU 4.
C

```

C OUTPUT

```

C     1 LIST OF VECTORS WITH OUTLIERS IDENTIFIED BY ASTERISKS
C     2 MEAN (ORIGINAL DATA)
C     3 STANDARD DEVIATION (ORIGINAL DATA)
C     4 MEAN (OUTLIERS DELETED)
C     5 STANDARD DEVIATION (OUTLIERS DELETED)

C
REAL*8 SUMX(10), XBAR(10), SD(10), XBAR1(10), SD1(10)
REAL*8 A(55), B(10), S(10,10)
REAL*8 RR, DELSTR, R(10), FMT(10), BLNK//      '//, VAR(10)
REAL*8 DEF(10)//      X1 ',', X2 ',', X3 ',', X4 ',', X5
1 ',', X6 ',', X7 ',', X8 ',', X9 ',', X10 '//'
REAL    X(501,10), Y(501,10)
INTEGER STAR//***//, BLANK//      '//, FLAG, FLGTOT, DSWI
INTEGER PREFMT(2)//'(A4,', 'T1, '//, OPT(2)//' YES',' NC'
INTEGER*2 IFLAG(500)
LOGICAL*1 FORMAT(89), RPAREN//')'
EQUIVALENCE (PREFMT,FORMAT(1)),(FMT,FCRMT(9)),(RPAREN,FORMAT(89))
1 DO 2 I=1,10
  VAR(I) = DEF(I)
2 FMT(I) = BLNK
  DO 5 I=1,500
5 IFLAG(I) = 0
  FLGTOT = 0
  LINES = 1
  NSWI = 0
  DSWI = 0
  OH = 0.01

```

APPENDIX B

```

IO = 1
IZ = 0
MM = 1
N = 1
READ(1,1000,END=999) FMT
READ(1,1010) NP, ALPH, ISWI
C      TEST IF NAMES HAVE BEEN ASSIGNED TO THE VARIABLES
IF(ISWI.GT.0) GO TO 10
MM = 2
GO TO 15
10 READ(1,1050) (VAR(I), I=1,NP)
C      READ IN THE DATA AS X-ARRAY
15 READ(1,FORMAT,END=995) IX, (X(N,I), I=1,NP)
16 IF(IXX.EQ.STAR) GO TO 19
N = N + 1
IF(N.GT.501) GO TO 990
GO TO 15
19 N = N - 1
C      SELECT APPROPRIATE F-VALUE
CALL FTABLE (N,NP,ALPH,FALPH)
C      WRITE THE INPUT CONTROL INFORMATION
WRITE(3,5002) NP, ALPH, FALPH, OPT(MM)
DELSTR = ((N-1)**2 * NP * FALPH) / (N * ((N-NP-1) + NP * FALPH))
20 DO 86 I=1,10
SUMX(I)=0.D0
DO 86 J=1,10
86 S(I,J)=0.D0
DO 30 J=1,N
C      TEST FOR FLAGGED VECTORS IDENTIFIED AS OUTLIERS
IF(IFLAG(J).GT.0) GO TO 30
DO 25 I=1,NP
25 SUMX(I) = SUMX(I) + X(J,I)
30 CONTINUE
C      FIND MEAN OF EACH VARIABLE
DO 40 I=1,NP
XBAR(I) = SUMX(I) / (N-FLG TOT)
DO 40 J=1,N
Y(J,I) = X(J,I) - XBAR(I)
40 IF(IFLAG(J).GT.0) Y(J,I) = C.D0
JJ = 0
DO 70 I=1,NP
DO 70 K=1,I
DO 60 J=1,N
60 S(I,K) = Y(J,I) * Y(J,K) + S(I,K)
C      FIND STANDARD DEVIATION OF EACH VARIABLE
SD(I) = DSQRT( S(I,I) / (N-1-FLG TOT))
JJ = JJ + 1
70 A(JJ) = S(I,K)
C      IF COMPUTATIONS ARE COMPLETE, BRANCH TO
C      PRINT TABLE OF MEANS AND S.D.'S
IF(DSWI.GT.0) GO TO 20C
CALL DSINV(A,VP,OH,IER)

```

APPENDIX B

```

      IF(IER) 991,80,991
C           WRITE THE LIST HEADING
 80  WRITE(3,1015) (VAR(I), I=1,NP)
    DO 100 J=1,N
    DO 90 K=1,NP
 90  B(K) = Y(J,K)
    CALL MPRD (B,A,R,IO,NP,IZ,IC,NP)
    CALL MPRD (R,B,RR,IO,NP,IZ,IZ,IO)
    RR = RR*(N-1)
    FLAG = BLANK
C           FLAG THIS VECTOR WITH AN ASTERISK IF
C           IT IS IDENTIFIED AS AN OUTLIER
    IF(RR.GT.DELSTR) FLAG = STAR
    IF(FLAG.NE.STAR) GO TO 95
    IFLAG(J) = 1
    FLGTOT = FLGTOT + 1
C           WRITE THE DATA VECTOR AND IF IDENTIFIED AS
C           AN OUTLIER, LABEL WITH AN ASTERISK
 95  WRITE(3,1020) J, RR, FLAG, (X(J,K), K=1,NP)
    LINES = LINES + 1
    IF(LINES.LE.55) GO TO 100
    WRITE(3,1015) (VAR(I), I=1,NP)
    LINES = 1
100 CONTINUE
    WRITE(3,1025) ALPH
    WRITE(3,1030) N
    WRITE(3,1035) FLGTOT
    WRITE(3,1040) DELSTR
C           SAVE MEANS AND S.D.'S, THEN CCP BACK AND
C           COMPUTE NEW MEANS AND S.D.'S AFTER DELETING OUTLIERS
    DO 110 I=1,NP
    XBAR1(I) = XBAR(I)
110 SD1(I) = SD(I)
    NSWI = 1
    GO TO 20
C           WRITE TABLE OF MEANS AND S.D.'S BEFORE AND
C           AFTER DELETION OF OUTLIERS
200 WRITE(3,2000)
    WRITE(3,2005)
    WRITE(3,2010)
    DO 210 I=1,NP
210 WRITE(3,2015) VAR(I), XBAR1(I), SD1(I), XBAR(I), SD(I)
    IF(NSWI.EQ.1) GO TO 999
    GO TO 1
C
990 WRITE (3,5000)
    GO TO 9999
991 WRITE (3,5001)
    GO TO 9999
995 NSWI = 1
    GO TO 19
999 WRITE (3,5009)

```

APPENDIX B

```
1000 FORMAT (20A4)
1010 FORMAT (I2,1X,F3.2,3X,I1)
1015 FORMAT (1H1,'LIST OF VECTORS WITH OUTLIERS IDENTIFIED BY ASTERISKS
1'//1H0,' J',7X,'DELTA',7X,1C(2X,A8)/)
1020 FORMAT (1H ,I3,F12.4,2X,A1,4X,10F10.2)
1025 FORMAT (///1H0,12X,'* OUTLIER IDENTIFIED AT ',F4.2,' SIGNIFICANCE
1LEVEL')
1030 FORMAT (1H ,12X,'SAMPLE SIZE IS 'I3)
1035 FORMAT (1H ,12X,'NO OF OUTLIERS IS ',I3)
1040 FORMAT (1H ,12X,'DELSTAR = 'F10.4)
1050 FORMAT (10A8)
2000 FORMAT (1H1,20X,'MEAN AND STANDARD DEVIATION OF THE VARIABLES')
2005 FORMAT (1H0,14X,'DATA BEFORE IDENTIFICATION',9X,'DATA AFTER DELETION'/1H ,22X,'OF OUTLIERS',21X,'OF OUTLIERS')
2010 FORMAT(1H0,'VARIABLES',8X,'MEAN',13X,'S.D.',11X,'MEAN',14X,'S.D.')
2015 FORMAT (1H0,A8,F15.4,F17.4,F15.4,F18.4)
5000 FORMAT (1H1,'SAMPLE SIZE EXCEEDS 500 - PROGRAM TERMINATED')
5001 FORMAT (1H1,'ERROR IN THE MATRIX INVERSION PROCESS - PROGRAM TERMINATED')
5002 FORMAT (1H1,'**INPUT CONTROL INFORMATION'//1H0,'**NUMBER OF VARIABLES IS',I4/1H , '**SIGNIFICANCE LEVEL IS',F5.2/1H , '**F-VALUE IS',
2F16.4//1H , '**VARIABLE NAME CARD: ',A4)
5009 FORMAT (1H1,'END OF JOB')
9999 STOP
END
```

APPENDIX B

```

SUBROUTINE FTABLE (N,M,SL,F)                                FT AB0000
C*****FTAB0000*****FTAB0010
C                                              * FT AB0020
C      FTABLE                                         * FT AB0030
C                                              * FT AB0040
C      SUBROUTINE FTABLE SELCTS THE PROPER VALUE OF F AT M AND N-M-1   * FT AB0050
C      DEGREES OF FREEDOM FOR SIGNIFICANCE LEVELS OF 5% OR 1%.        * FT AB0060
C                                              * FT AB0070
C      CALLING PARAMETERS                                     * FT AB0080
C                                              * FT AB0090
C      N = SAMPLE SIZE                                     INTEGER          * FT AB0100
C      M = NUMBER OF VARIABLES                            M.LE.10          * FT AB0110
C      SL = SIGNIFICANCE LEVEL                           .05 CR .01          REAL           * FT AB0120
C      F = SELECTED F-VALUE                             REAL            * FT AB0130
C                                              * FT AB0140
C*****FTAB0150*****FTAB0160
DIMENSION TABLE(2,10,37)                                FT AB0160
INTEGER XDF(7)/30,40,60,120,200,400,1000/, DF          FT AB0170
REAL SL T(2)/.05,.01/                                    FT AB0180
REAL TABL1(200)/161.4,4052.,199.5,4999.5,215.7,5403.,224.6,5625., FT AB0190
1230.2,5764.,234.0,5859.,236.8,5928.,238.9,5982.,240.5,6022.,241.9, FT AB0200
26056.,18.51,98.50,19.00,99.00,19.16,99.17,19.25,99.25,19.30,99.30, FT AB0210
319.33,99.33,19.35,99.36,19.37,99.37,19.38,99.39,19.40,99.40,10.13, FT AB0220
434.12, 9.55,30.82, 9.28,29.46, 9.12,28.71, 9.01,28.24, 8.94,27.91, FT AB0230
5 8.89,27.67, 8.85,27.49, 8.81,27.35, 8.79,27.23, 7.71,21.20, 6.94, FT AB0240
618.00, 6.59,16.69, 6.39,15.98, 6.26,15.52, 6.16,15.21, 6.09,14.98, FT AB0250
7 6.04,14.80, 6.00,14.66, 5.96,14.55, 6.61,16.26, 5.79,13.27, 5.41, FT AB0260
812.06, 5.19,11.39, 5.05,10.97, 4.95,10.67, 4.88,10.46, 4.82,10.29, FT AB0270
9 4.77,10.16, 4.74,10.05, 5.99,13.75, 5.14,10.92, 4.76, 9.78, 4.53, FT AB0280
1 9.15, 4.39, 8.75, 4.28, 8.47, 4.21, 8.26, 4.15, 8.10, 4.10, 7.98, FT AB0290
2 4.06, 7.87, 5.59,12.25, 4.74, 9.55, 4.35, 8.45, 4.12, 7.85, 3.97, FT AB0300
3 7.46, 3.87, 7.19, 3.79, 6.99, 3.73, 6.84, 3.68, 6.72, 3.64, 6.62, FT AB0310
4 5.32,11.26, 4.46, 8.65, 4.07, 7.59, 3.84, 7.01, 3.69, 6.63, 3.58, FT AB0320
5 6.37, 3.50, 6.18, 3.44, 6.03, 3.39, 5.91, 3.35, 5.81, 5.12,10.56, FT AB0330
6 4.26, 8.02, 3.86, 6.99, 3.63, 6.42, 3.48, 6.05, 3.37, 5.80, 3.29, FT AB0340
7 5.61, 3.23, 5.47, 3.18, 5.35, 3.14, 5.26, 4.96,10.04, 4.10, 7.56, FT AB0350
8 3.71, 6.55, 3.48, 5.99, 3.33, 5.64, 3.22, 5.39, 3.14, 5.20, 3.07, FT AB0360
9 5.06, 3.02, 4.94, 2.98, 4.85/                               FT AB0370
REAL TABL2(200)/ 4.84,9.65,3.98,7.21,3.59,6.22,3.36,5.67,3.20, FT AB0380
15.32,3.09,5.07,3.01,4.89,2.95,4.74,2.90,4.63,2.85,4.54,4.75,9.33, FT AB0390
23.89,6.93,3.49,5.95,3.26,5.41,3.11,5.06,3.00,4.82,2.91,4.64,2.85, FT AB0400
34.50,2.80,4.39,2.75,4.30,4.67,9.07,3.81,6.70,3.41,5.74,3.18,5.21, FT AB0410
43.03,4.86,2.92,4.62,2.83,4.44,2.77,4.30,2.71,4.19,2.67,4.10,4.60, FT AB0420
58.86,3.74,6.51,3.34,5.56,3.11,5.04,2.96,4.69,2.85,4.46,2.76,4.28, FT AB0430
62.70,4.14,2.65,4.03,2.60,3.94,4.54,8.68,3.68,6.36,3.29,5.42,3.06, FT AB0440
74.89,2.90,4.56,2.75,4.32,2.71,4.14,2.64,4.00,2.59,3.89,2.54,3.80, FT AB0450
84.49,8.53,3.63,6.23,3.24,5.29,3.01,4.77,2.85,4.44,2.74,4.20,2.66, FT AB0460
94.03,2.59,3.89,2.54,3.78,2.49,3.69,4.45,8.40,3.59,6.11,3.20,5.18, FT AB0470
12.96,4.67,2.81,4.34,2.70,4.10,2.61,3.93,2.55,3.79,2.49,3.68,2.45, FT AB0480
23.59,4.41,8.29,3.55,6.01,3.16,5.09,2.93,4.58,2.77,4.25,2.66,4.01, FT AB0490
32.58,3.84,2.51,3.71,2.46,3.60,2.41,3.51,4.38,8.18,3.52,5.93,3.13, FT AB0500
45.01,2.90,4.50,2.74,4.17,2.63,3.94,2.54,3.77,2.48,3.63,2.42,3.52, FT AB0510

```

APPENDIX B

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52.38,3.43,4.35,8.10,3.49,5.85,3.10,4.94,2.87,4.43,2.71,4.10,2.60, FT AB0520
63.87,2.51,3.70,2.45,3.56,2.39,3.46,2.35,3.37/ FT AB0530
REAL TABL3(2001)/ 4.32,8.02,3.47,5.78,3.07,4.87,2.84,4.37,2.68, FT AB0540
14.04,2.57,3.81,2.49,3.64,2.42,3.51,2.37,3.40,2.32,3.31,4.30,7.95, FT AB0550
23.44,5.72,3.05,4.82,2.82,4.31,2.66,3.99,2.55,3.76,2.46,3.59,2.40, FT AB0560
33.45,2.34,3.35,2.30,3.26,4.28,7.88,3.42,5.66,3.03,4.76,2.80,4.26, FT AB0570
42.64,3.94,2.53,3.71,2.44,3.54,2.37,3.41,2.32,3.30,2.27,3.21,4.26, FT AB0580
57.82,3.40,5.61,3.01,4.72,2.78,4.22,2.62,3.90,2.51,3.67,2.42,3.50, FT AB0590
62.36,3.36,2.30,3.26,2.25,3.17,4.24,7.77,3.39,5.57,2.99,4.68,2.76, FT AB0600
74.18,2.60,3.85,2.49,3.63,2.40,3.46,2.34,3.32,2.28,3.22,2.24,3.13, FT AB0610
84.23,7.72,3.37,5.53,2.98,4.64,2.74,4.14,2.59,3.82,2.47,3.59,2.39, FT AB0620
93.42,2.32,3.29,2.27,3.18,2.22,3.09,4.21,7.68,3.35,5.49,2.96,4.60, FT AB0630
12.73,4.11,2.57,3.78,2.46,3.56,2.37,3.39,2.31,3.26,2.25,3.15,2.20, FT AB0640
23.06,4.20,7.64,3.34,5.45,2.95,4.57,2.71,4.07,2.56,3.75,2.45,3.53, FT AB0650
32.36,3.36,2.29,3.23,2.24,3.12,2.19,3.03,4.18,7.60,3.33,5.42,2.93, FT AB0660
44.54,2.70,4.04,2.55,3.73,2.43,3.50,2.35,3.33,2.28,3.20,2.22,3.09, FT AB0670
52.18,3.00,4.17,7.56,3.32,5.39,2.92,4.51,2.69,4.02,2.53,3.70,2.42, FT AB0680
63.47,2.33,3.30,2.27,3.17,2.21,3.07,2.16,2.98/ FT AB0690
REAL TABL4(1401)/ 4.08,7.31,3.23,5.18,2.84,4.31,2.61,3.83,2.45, FT AB0700
13.51,2.34,3.29,2.25,3.12,2.18,2.99,2.12,2.89,2.08,2.80,4.00,7.08, FT AB0710
23.15,4.98,2.76,4.13,2.53,3.65,2.37,3.34,2.25,3.12,2.17,2.95,2.10, FT AB0720
32.82,2.04,2.72,1.99,2.63,3.92,6.85,3.07,4.79,2.68,3.95,2.45,3.48, FT AB0730
42.29,3.17,2.17,2.96,2.09,2.79,2.02,2.66,1.96,2.56,1.91,2.47,3.89, FT AB0740
56.76,3.04,4.71,2.65,3.88,2.41,3.41,2.26,3.11,2.14,2.90,2.05,2.73, FT AB0750
61.98,2.60,1.92,2.50,1.87,2.41,3.86,6.70,3.02,4.66,2.62,3.83,2.39, FT AB0760
73.36,2.23,3.06,2.12,2.85,2.03,2.69,1.96,2.55,1.90,2.46,1.85,2.37, FT AB0770
83.85,6.66,3.00,4.62,2.61,3.80,2.38,3.34,2.22,3.04,2.10,2.82,2.02, FT AB0780
92.66,1.95,2.53,1.89,2.43,1.84,2.34,3.84,6.63,3.00,4.61,2.60,3.78, FT AB0790
12.37,3.32,2.21,3.02,2.10,2.80,2.01,2.64,1.94,2.51,1.88,2.41,1.83, FT AB0800
22.32/ FT AB0810
EQUIVALENCE (TABLE(1,1,1),TABL1(1)), (TABLE(1,1,11),TABL2(1)) FT AB0820
EQUIVALENCE (TABLE(1,1,21),TABL3(1)), (TABLE(1,1,31),TABL4(1)) FT AB0830
IF(M.GT.10) GO TO 590 FT AB0840
DF = N-M-1 FT AB0850
IF(SL.EQ.SLT(1)) GO TO 10 FT AB0860
IF(SL.EQ.SLT(2)) GO TO 15 FT AB0870
GO TO 591 FT AB0880
10 L = 1 FT AB0890
GO TO 16 FT AB0900
15 L = 2 FT AB0910
16 IF(DF.LE.30) GO TO 40 FT AB0920
DO 20 I=2,7 FT AB0930
IF(DF-XDF(I)) 50,30,20 FT AB0940
20 CONTINUE FT AB0950
F = TABLE(L,M,36) + ((1./N)/.001) * (TABLE(L,M,36)-TABLE(L,M,37)) FT AB0960
RETURN FT AB0970
30 DF = I+29 FT AB0980
40 F = TABLE(L,M,DF) FT AB0990
RETURN FT AB1000
50 F = TABLE(L,M,I+29) + ((1./DF - 1./XDF(I)) / (1./XDF(I-1) - 1./XDF FT AB1010
I(I))) * (TABLE(L,M,I+28) - TABLE(L,M,I+29)) FT AB1020
RETURN FT AB1030
FT AB1040
590 WRITE(3,990) FT AB1050
F = 2.32 FT AB1060
RETURN FT AB1070
591 WRITE(3,991) SL FT AB1080
SL = 0.01 FT AB1090
GO TO 15 FT AB1100
990 FORMAT (1H1,'**NUMBER OF VARIABLES > 10. F HAS BEEN SET TO A DUMMY') FT AB1110
1Y VALUE OF 2.32') FT AB1120
991 FORMAT (1H1,'**SIGNIFICANCE LEVEL ',F6.3,' IS NOT ACCEPTABLE. LEVEL FT AB1130
IEL IS SET TO .01 .') FT AB1140
END

```

APPENDIX B

```

C ..... MPRD 001
C ..... MPRD 002
C ..... MPRD 003
C ..... MPRD 004
C ..... MPRD 005
C ..... MPRD 006
C ..... MPRD 007
C ..... MPRD 008
C ..... MPRD 009
C ..... MPRD 010
C ..... MPRD 011
C ..... MPRD 012
C ..... MPRD 013
C ..... MPRD 014
C ..... MPRD 015
C ..... MPRD 016
C ..... MPRD 017
C ..... MPRD 018
C ..... MPRD 019
C ..... MPRD 020
C ..... MPRD 021
C ..... MPRD 022
C ..... MPRD 023
C ..... MPRD 024
C ..... MPRD 025
C ..... MPRD 026
C ..... MPRD 027
C ..... MPRD 028
C ..... MPRD 029
C ..... MPRD 030
C ..... MPRD 031
C ..... MPRD 032
C ..... MPRD 033
C ..... MPRD 034
C ..... MPRD 035
C ..... MPRD 036
C ..... MPRD 037
C ..... MPRD 038
C ..... MPRD 039
C ..... MPRD 040
C ..... MPRD 041
C ..... MPRD 042
C ..... MPRD 043
C ..... MPRD 044
C ..... MPRD 045
C ..... MPRD 046
C ..... MPRD 047
C ..... MPRD 048
C ..... MPRD 049
C ..... MPRD 050
C ..... MPRD 051
C ..... MPRD 052

C SUBROUTINE MPRD
C
C PURPOSE
C   MULTIPLY TWO MATRICES TO FORM A RESULTANT MATRIX
C
C USAGE
C   CALL MPRD(A,B,R,N,M,MSA,MSB,L)
C
C DESCRIPTION OF PARAMETERS
C   A - NAME OF FIRST INPUT MATRIX
C   B - NAME OF SECOND INPUT MATRIX
C   R - NAME OF OUTPUT MATRIX
C   N - NUMBER OF ROWS IN A AND R
C   M - NUMBER OF COLUMNS IN A AND ROWS IN B
C   MSA - ONE DIGIT NUMBER FOR STORAGE MODE OF MATRIX A
C         0 - GENERAL
C         1 - SYMMETRIC
C         2 - DIAGONAL
C   MSB - SAME AS MSA EXCEPT FOR MATRIX B
C   L - NUMBER OF COLUMNS IN B AND R
C
C REMARKS
C   MATRIX R CANNOT BE IN THE SAME LOCATION AS MATRICES A OR B
C   NUMBER OF COLUMNS OF MATRIX A MUST BE EQUAL TO NUMBER OF ROWS
C   OF MATRIX B
C
C SUBROUTINES AND FUNCTION SUBPROGRAMS REQUIRED
C   LOC
C
C METHOD
C   THE M BY L MATRIX B IS PREMULTIPLIED BY THE N BY M MATRIX A
C   AND THE RESULT IS STORED IN THE N BY L MATRIX R. THIS IS A
C   ROW INTO COLUMN PRODUCT.
C   THE FOLLOWING TABLE SHOWS THE STORAGE MODE OF THE OUTPUT
C   MATRIX FOR ALL COMBINATIONS OF INPUT MATRICES
C
C      A          B          R          MPRD
C      GENERAL    GENERAL    GENERAL    039
C      GENERAL    SYMMETRIC  GENERAL    040
C      GENERAL    DIAGONAL  GENERAL    041
C      SYMMETRIC  GENERAL    GENERAL    042
C      SYMMETRIC  SYMMETRIC  GENERAL    043
C      SYMMETRIC  DIAGONAL  GENERAL    044
C      DIAGONAL   GENERAL    GENERAL    045
C      DIAGONAL   SYMMETRIC  GENERAL    046
C      DIAGONAL   DIAGONAL   DIAGONAL  047
C
C
C SUBROUTINE MPRD(A,B,R,N,M,MSA,MSB,L)

```

APPENDIX B

```

DOUBLE PRECISION A, B, R
DIMENSION A(1),B(1),R(1)                                MPRD 053
C
C      SPECIAL CASE FOR DIAGONAL BY DIAGONAL             MPRD 054
C
C      MS=MSA*10+MSB                                     MPRD 055
C      IF(MS-22) 30,1C,30                                 MPRD 056
10 DO 20 I=1,N                                         MPRD 057
20 R(I)=A(I)*B(I)                                     MPRD 058
      RETURN                                              MPRD 059
C
C      ALL OTHER CASES                                  MPRD 060
C
30 IR=1                                                 MPRD 061
DO 90 K=1,L                                         MPRD 062
DO 90 J=1,N                                         MPRD 063
R(IR)=0                                               MPRD 064
DO 80 I=1,M                                         MPRD 065
IF(MS) 40,60,4C                                     MPRD 066
40 CALL LOC(J,I,IA,N,M,MSA)                         MPRD 067
      CALL LOC(I,K,IB,M,L,MSB)
      IF(IA) 50,8C,5C
50 IF(IB) 70,80,7C
60 IA=N*(I-1)+J                                     MPRD 068
      IB=M*(K-1)+I
70 R(IR)=R(IR)+A(IA)*B(IB)                         MPRD 069
80 CONTINUE                                           MPRD 070
90 IR=IR+1                                         MPRD 071
      RETURN                                              MPRD 072
      END                                                 MPRD 073
                                                MPRD 074
                                                MPRD 075
                                                MPRD 076
                                                MPRD 077
                                                MPRD 078
                                                MPRD 079
                                                MPRD 080
                                                MPRD 081

```

APPENDIX B

```

C                               LOC 001
C .....LOC 002
C                               LOC 003
C   SUBROUTINE LOC               LOC 004
C                               LOC 005
C   PURPOSE                   LOC 006
C     COMPUTE A VECTOR SUBSCRIPT FOR AN ELEMENT IN A MATRIX OF LOC 007
C     SPECIFIED STORAGE MODE   LOC 008
C                               LOC 009
C   USAGE                     LOC 010
C     CALL LOC %I,J,IR,N,M,MS< LOC 011
C                               LOC 012
C   DESCRIPTION OF PARAMETERS  LOC 013
C     I - ROW NUMBER OF ELEMENT LOC 014
C     J - COLUMN NUMBER OF ELEMENT LOC 015
C     IR - RESULTANT VECTOR SUBSCRIPT LOC 016
C     N - NUMBER OF ROWS IN MATRIX LOC 017
C     M - NUMBER OF COLUMNS IN MATRIX LOC 018
C     MS - ONE DIGIT NUMBER FOR STORAGE MODE OF MATRIX1 LOC 019
C       0 - GENERAL             LOC 020
C       1 - SYMMETRIC            LOC 021
C       2 - DIAGONAL             LOC 022
C                               LOC 023
C   REMARKS                  LOC 024
C     NONE                     LOC 025
C                               LOC 026
C   SUBROUTINES AND FUNCTION SUBPROGRAMS REQUIRED LOC 027
C     NONE                     LOC 028
C                               LOC 029
C   METHOD                    LOC 030
C     MS#0  SUBSCRIPT IS COMPUTED FOR A MATRIX WITH N*M ELEMENTS LOC 031
C           IN STORAGE %GENERAL MATRIX<
C     MS#1  SUBSCRIPT IS COMPUTED FOR A MATRIX WITH N*N<1</2 IN LOC 032
C           STORAGE %UPPER TRIANGLE OF SYMMETRIC MATRIX<. IF LOC 033
C           ELEMENT IS IN LOWER TRIANGULAR PORTION, SUBSCRIPT IS LOC 034
C           CORRESPONDING ELEMENT IN UPPER TRIANGLE. LOC 035
C     MS#2  SUBSCRIPT IS COMPUTED FOR A MATRIX WITH N ELEMENTS LOC 037
C           IN STORAGE %DIAGONAL ELEMENTS OF DIAGONAL MATRIX<. LOC 038
C           IF ELEMENT IS NOT ON DIAGONAL %AND THEREFORE NOT IN LOC 039
C           STORAGE<, IR IS SET TO ZERO. LOC 040
C                               LOC 041
C .....LOC 042
C                               LOC 043
C   SUBROUTINE LOC(I,J,IR,N,M,MS) LOC 044
C                               LOC 045
C     IX=I                      LOC 046
C     JX=J                      LOC 047
C     IF(MS-1) 10,20,3C          LOC 048
10    IRX=N*(JX-1)+IX          LOC 049
      GO TO 36                  LOC 050
20    IF(IX-JX) 22,24,24        LOC 051
22    IRX=IX+(JX*JX-IX)/2      LOC 052
      GO TO 36                  LOC 053
24    IRX=JX+(IX*IX-IX)/2      LOC 054
      GO TO 36                  LOC 055
30    IRX=0                      LOC 056
      IF(IX-JX) 36,32,36        LOC 057
32    IRX=IX                      LOC 058
36    IR=IRX                      LOC 059
      RETURN                      LOC 060
      END                         LOC 061

```

APPENDIX B

		S INV	10
.....		S INV	20
		S INV	30
SUBROUTINE DSINV			
PURPOSE		S INV	50
INVERT A GIVEN SYMMETRIC POSITIVE DEFINITE MATRIX		S INV	60
USAGE		S INV	70
CALL DSINV(A,N,EPS,IER)		S INV	80
DESCRIPTION OF PARAMETERS		S INV	90
A	- DOUBLE PRECISION UPPER TRIANGULAR PART OF GIVEN SYMMETRIC POSITIVE DEFINITE N BY N COEFFICIENT MATRIX.	S INV	110
	ON RETURN A CONTAINS THE RESULTANT UPPER TRIANGULAR MATRIX IN DOUBLE PRECISION.	S INV	120
N	- THE NUMBER OF ROWS (COLUMNS) IN GIVEN MATRIX.	S INV	150
EPS	- SINGLE PRECISION INPUT CONSTANT WHICH IS USED AS RELATIVE TOLERANCE FOR TEST ON LOSS OF SIGNIFICANCE.	S INV	170
IER	- RESULTING ERROR PARAMETER CODED AS FOLLOWS	S INV	200
	IER=0 - NO ERROR	S INV	210
	IER=-1 - NO RESULT BECAUSE OF WRONG INPUT PARAMETER N OR BECAUSE SOME RADICAND IS NON-POSITIVE (MATRIX A IS NOT POSITIVE DEFINITE, POSSIBLY DUE TO LOSS OF SIGNIFICANCE)	S INV	220
	IER=K - WARNING WHICH INDICATES LOSS OF SIGNIFICANCE. THE RADICAND FORMED AT FACTORIZATION STEP K+1 WAS STILL POSITIVE BUT NO LONGER GREATER THAN ABS(EPS*A(K+1,K+1)).	S INV	230
		S INV	240
		S INV	250
		S INV	260
		S INV	270
		S INV	280
		S INV	290
		S INV	300
		S INV	310
REMARKS		S INV	320
	THE UPPER TRIANGULAR PART OF GIVEN MATRIX IS ASSUMED TO BE STORED COLUMNWISE IN N*(N+1)/2 SUCCESSIVE STORAGE LOCATIONS.	S INV	330
	IN THE SAME STORAGE LOCATIONS THE RESULTING UPPER TRIANGULAR MATRIX IS STORED COLUMNWISE TOO.	S INV	340
	THE PROCEDURE GIVES RESULTS IF N IS GREATER THAN 0 AND ALL CALCULATED RADICANDS ARE POSITIVE.	S INV	350
		S INV	360
		S INV	370
		S INV	380
		S INV	390
SUBROUTINES AND FUNCTION SUBPROGRAMS REQUIRED		S INV	400
DMFSD		S INV	420
METHOD		S INV	430
	SOLUTION IS DONE USING FACTORIZATION BY SUBROUTINE DMFSD.	S INV	450
.....		S INV	460
		S INV	470
SUBROUTINE DSINV(A,N,EPS,IER)		S INV	490
		S INV	500

APPENDIX B

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DIMENSION A(1)                                SINV 510
DOUBLE PRECISION A,DIN,WORK
C
C      FACTORIZE GIVEN MATRIX BY MEANS OF SUBROUTINE DMFSD    SINV 530
C      A = TRANSPSE(T) * T                                     SINV 550
C      CALL DMFSD(A,V,EPS,IER)
IF(IER) 9,1,1
C
C      INVERT UPPER TRIANGULAR MATRIX T                      SINV 570
C      PREPARE INVERSION-LOOP                                SINV 580
C
1 IP IV=N*(N+1)/2                            SINV 590
IND=IP IV
C
C      INITIALIZE INVERSION-LOOP                           SINV 600
DO 6 I=1,N
DIN=1.D0/A(IP IV)
A(IP IV)=DIN
MIN=N
KEND=I-1
LANF=N-KEND
IF(KEND) 5,5,2
2 J=IND
C
C      INITIALIZE ROW-LOOP                                 SINV 610
DO 4 K=1,KEND
WORK=0.D0
MIN=MIN-1
LHOR=IP IV
LVER=J
C
C      START INNER LOOP                                  SINV 620
DO 3 L=LANF,MIN
LVER=LVER+1
LHOR=LHOR+L
3 WORK=WORK+A(LVER)*A(LHOR)
      END OF INNER LOOP                               SINV 630
C
      A(J)=-WORK*DIN                                SINV 640
4 J=J-MIN
      END OF ROW-LOOP                                SINV 650
C
5 IP IV=IP IV-MIN                            SINV 670
6 IND=IND-1
      END OF INVERSION-LOOP                         SINV 680
C
C      CALCULATE INVERSE(A) BY MEANS OF INVERSE(T)        SINV 690
C      INVERSE(A) = INVERSE(T) * TRANSPOSE(INVERSE(T))   SINV 700
C      INITIALIZE MULTIPLICATION-LOOP                  SINV 710
DO 8 I=1,N
IP IV=IP IV+I
J=IP IV
C

```

APPENDIX B

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C      INITIALIZE ROW- LOOP           SINV 1030
DO 8 K=I,N                         SINV 1040
WORK=0.D0                           SINV 1050
LHOR=J                            SINV 1060
C
C      START INNER LOOP            SINV 1070
DO 7 L=K,N                         SINV 1080
LVER=LHOR+K-I                      SINV 1090
WORK=WORK+A(LHOR)*A(LVER)          SINV 1100
7 LHOR=LHOR+L                     SINV 1120
C      END OF INNER LOOP          SINV 1130
SINV 1140
C      A(J)=WORK                 SINV 1150
8 J=J+K                           SINV 1160
C      END OF ROW- AND MULTIPLICATION-LOOP
C
9 RETURN
END

```

APPENDIX B

```

C ..... MFSD 10
C ..... MFSD 20
C ..... MFSD 30
C SUBROUTINE DMFSD
C ..... MFSD 50
C PURPOSE MFSD 60
C FACTOR A GIVEN SYMMETRIC POSITIVE DEFINITE MATRIX MFSD 70
C ..... MFSD 80
C USAGE MFSD 90
C CALL DMFSD(A,N,EPS,IER) MFSD 110
C DESCRIPTION OF PARAMETERS MFSD 120
C A - DOUBLE PRECISION UPPER TRIANGULAR PART OF GIVEN MFSD 150
C SYMMETRIC POSITIVE DEFINITE N BY N COEFFICIENT
C MATRIX.
C ON RETURN A CONTAINS THE RESULTANT UPPER MFSD 150
C TRIANGULAR MATRIX IN DOUBLE PRECISION.
C N - THE NUMBER OF ROWS (COLUMNS) IN GIVEN MATRIX. MFSD 170
C EPS - SINGLE PRECISION INPUT CONSTANT WHICH IS USED
C AS RELATIVE TOLERANCE FOR TEST ON LOSS OF
C SIGNIFICANCE.
C IER - RESULTING ERROR PARAMETER CODED AS FOLLOWS MFSD 200
C IER=0 - NO ERROR MFSD 210
C IER=-1 - NO RESULT BECAUSE OF WRONG INPUT PARAMETER N OR BECAUSE MFSD 220
C RADICAND IS NON-POSITIVE (MATRIX A IS NOT POSITIVE MFSD 230
C DEFINITE, POSSIBLY DUE TO LOSS OF SIGNIFICANCE) MFSD 240
C IER=K - WARNING WHICH INDICATES LOSS OF SIGNIFICANCE. THE RADICAND FORMED AT FACTORIZATION MFSD 250
C STEP K+1 WAS STILL POSITIVE BUT NO LONGER GREATER THAN ABS(EPS*A(K+1,K+1)). MFSD 260
C MFSD 270
C MFSD 280
C MFSD 290
C MFSD 300
C MFSD 310
C REMARKS MFSD 320
C THE UPPER TRIANGULAR PART OF GIVEN MATRIX IS ASSUMED TO BE MFSD 330
C STORED COLUMNWISE IN N*(N+1)/2 SUCCESSIVE STORAGE LOCATIONS. MFSD 340
C IN THE SAME STORAGE LOCATIONS THE RESULTING UPPER TRIANGULAR MATRIX IS STORED COLUMNWISE TOO. MFSD 350
C MFSD 360
C THE PROCEDURE GIVES RESULTS IF N IS GREATER THAN 0 AND ALL MFSD 370
C CALCULATED RADICANDS ARE POSITIVE. MFSD 380
C THE PRODUCT OF RETURNED DIAGONAL TERMS IS EQUAL TO THE MFSD 390
C SQUARE-ROOT OF THE DETERMINANT OF THE GIVEN MATRIX. MFSD 400
C MFSD 410
C SUBROUTINES AND FUNCTION SUBPROGRAMS REQUIRED MFSD 420
C NONE MFSD 430
C ..... MFSD 440
C METHOD MFSD 450
C SOLUTION IS DONE USING THE SQUARE-ROOT METHOD OF CHOLESKY. MFSD 460
C THE GIVEN MATRIX IS REPRESENTED AS PRODUCT OF TWO TRIANGULAR MFSD 470
C MATRICES, WHERE THE LEFT HAND FACTOR IS THE TRANSPOSE OF MFSD 480
C THE RETURNED RIGHT HAND FACTOR. MFSD 490
C ..... MFSD 510

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APPENDIX B

```

C          MFSD 500
C          MFSD 520
C
C      SUBROUTINE DMFSD(A,N,EPS,IER)
C
C          MFSD 540
C          MFSD 550
C          MFSD 560
C
C      DIMENSION A(1)
C      DOUBLE PRECISION DPIV,DSUM,A
C
C          MFSD 580
C          MFSD 590
C          MFSD 600
C          MFSD 610
C          MFSD 620
C          MFSD 630
C          MFSD 640
C          MFSD 650
C          MFSD 660
C          MFSD 670
C          MFSD 680
C
C          MFSD 690
C          MFSD 700
C
C      TEST ON WRONG INPUT PARAMETER N
C      IF(N-1) 12,1,1
C      1 IER=0
C
C      INITIALIZE DIAGONAL-LOOP
C      KPIV=0
C      DO 11 K=1,N
C      KPIV=KPIV+K
C      IND=KPIV
C      LEND=K-1
C
C          MFSD 720
C          MFSD 730
C          MFSD 740
C          MFSD 750
C          MFSD 760
C          MFSD 770
C          MFSD 780
C          MFSD 790
C          MFSD 800
C          MFSD 810
C
C      CALCULATE TOLERANCE
C      TOL=ABS(EPS*SNGL(A(KPIV)))
C
C      START FACTORIZATION-LOOP OVER K-TH ROW
C      DO 11 I=K,N
C      DSUM=0.D0
C      IF(LEND) 2,4,2
C
C          MFSD 830
C          MFSD 840
C          MFSD 850
C
C      START INNER LOOP
C      2 DO 3 L=1,LEND
C      LANF=KPIV-L
C      LIND=IND-L
C      3 DSUM=DSUM+A(LANF)*A(LIND)
C          END OF INNER LOOP
C
C          MFSD 870
C          MFSD 880
C          MFSD 890
C          MFSD 900
C          MFSD 910
C          MFSD 920
C          MFSD 930
C          MFSD 940
C          MFSD 950
C          MFSD 960
C          MFSD 970
C          MFSD 980
C          MFSD 990
C
C          MFSD 1000
C          MFSD 1010
C          MFSD 1020
C          MFSD 1030
C          MFSD 1040
C          MFSD 1050
C          MFSD 1060
C          MFSD 1070
C          MFSD 1080
C          MFSD 1090
C
C      COMPUTE PIVOT ELEMENT
C      9 DPIV=DSQRT(DSUM)
C      A(KPIV)=DPIV
C      DPIV=1.D0/DPIV
C      GO TO 11
C
C          MFSD 1100
C          MFSD 1110
C          MFSD 1120
C          MFSD 1130
C          MFSD 1140
C          MFSD 1150
C          MFSD 1160
C          MFSD 1170
C          MFSD 1180
C          MFSD 1190
C
C      CALCULATE TERMS IN ROW
C      10 A(IND)=DSUM*DPIV
C      11 IND=IND+1
C
C          MFSD 1200
C          MFSD 1210
C          MFSD 1220
C          MFSD 1230
C          MFSD 1240
C          MFSD 1250
C          MFSD 1260
C          MFSD 1270
C          MFSD 1280
C          MFSD 1290
C
C      END OF DIAGONAL-LOOP
C      RETURN
C      12 IER=-1
C      RETURN
C      END

```

APPENDIX C

OUTPUT FOR THE EXAMPLE

LIST OF VECTORS WITH OUTLIERS IDENTIFIED BY ASTERISKS

J	DELTA	H/R	J	DELTA	H/R
1	0.6901	91.12	34	1.1316	86.42
2	0.0240	96.05	35	0.8436	90.48
3	0.0139	96.32	36	2.7647	85.05
4	0.4924	102.30	37	0.0249	96.03
5	0.3994	92.57	38	0.0203	98.22
6	0.0446	95.64	39	0.0452	95.63
7	0.6192	91.44	40	1.5831	106.36
8	0.3098	93.12	41	0.0170	96.23
9	0.1806	94.08	42	0.7866	90.71
10	0.5239	91.90	43	0.1973	93.94
11	0.0063	96.60	44	0.2601	93.46
12	0.0173	98.14	45	0.2643	93.43
13	0.3266	101.35	46	0.0440	95.65
14	0.1055	99.55	47	0.0087	96.50
15	0.0390	95.74	48	0.4256	101.94
16	0.0244	98.32	49	2.1912	107.98
17	0.1420	99.93	50	0.0968	99.45
18	0.6830	103.21	51	3.7721	111.35
19	0.5996	102.83	52	0.9550	104.31
20	0.1279	99.79	53	0.7815	103.63
21	1.1196	104.90	54	0.3489	101.49
22	0.0348	98.58	55	3.8632	111.52
23	0.6898	103.24	56	7.8031 *	117.56
24	0.9047	104.12	57	3.2634	110.36
25	0.2784	101.03	58	0.0253	98.34
26	0.8715	90.37	59	0.0146	96.30
27	0.7036	103.30	60	4.2446	82.15
28	1.8903	87.15	61	3.9288	82.72
29	2.6336	105.02	62	3.2739	83.98
30	0.0016	96.89	63	0.7131	91.02
31	0.5893	91.58	64	0.1841	94.05
32	2.7238	85.14	65	0.9473	90.08
33	1.9017	87.12	66	0.4621	102.14

* OUTLIER IDENTIFIED AT 0.01 SIGNIFICANCE LEVEL

SAMPLE SIZE IS 66

NO OF OUTLIERS IS 1

DELSTAR = 6.3530

DATA BEFORE IDENTIFICATION OF OUTLIERS			DATA AFTER DELETION OF OUTLIERS		
VARIABLES	MEAN	S.D.	MEAN	S.D.	
H/R	97.1806	7.2956	96.8671	6.8897	

APPENDIX C

LIST OF VECTORS WITH OUTLIERS IDENTIFIED BY ASTERISKS

J	DELTA	G	J	DELTA	G
1	1.3790	0.70	34	0.8358	0.85
2	1.1829	0.75	35	0.6848	0.90
3	1.0018	0.80	36	0.5488	0.95
4	0.8358	0.85	37	0.4279	1.00
5	0.6848	0.90	38	0.3220	1.05
6	0.5488	0.95	39	0.2311	1.10
7	0.4279	1.00	40	0.1553	1.15
8	0.3220	1.05	41	0.0945	1.20
9	0.2311	1.10	42	0.0487	1.25
10	0.1553	1.15	43	0.0180	1.30
11	0.0945	1.20	44	0.0022	1.35
12	0.0487	1.25	45	0.0016	1.40
13	0.0180	1.30	46	0.0159	1.45
14	0.0022	1.35	47	0.0897	1.55
15	0.0016	1.40	48	0.1491	1.60
16	0.0159	1.45	49	0.2236	1.65
17	0.0453	1.50	50	0.3131	1.70
18	0.0897	1.55	51	0.4177	1.75
19	0.1491	1.60	52	0.5372	1.80
20	0.3131	1.70	53	0.9861	1.95
21	0.5372	1.80	54	1.7951	2.15
22	0.6718	1.85	55	2.0349	2.20
23	0.8215	1.90	56	4.1346	2.55
24	0.9861	1.95	57	4.4947	2.60
25	1.5703	2.10	58	14.1922 *	3.55
26	1.7951	2.15	59	1.0018	0.80
27	2.0349	2.20	60	0.8358	0.85
28	2.3137	2.50	61	0.6848	0.90
29	2.0575	0.55	62	0.5488	0.95
30	1.5901	0.65	63	0.4279	1.00
31	1.3790	0.70	64	0.3220	1.05
32	1.1829	0.75	65	0.0897	1.55
33	1.0018	0.80	66	2.8446	2.35

* OUTLIER IDENTIFIED AT 0.01 SIGNIFICANCE LEVEL

SAMPLE SIZE IS 66

NO OF OUTLIERS IS 1

DELSTAR = 6.3530

DATA BEFORE IDENTIFICATION OF OUTLIERS

VARIABLES

MEAN

S.D.

G

1.3773

0.5767

DATA AFTER DELETION OF OUTLIERS

MEAN

1.3438

S.D.

0.5128

APPENDIX C

LIST OF VECTORS WITH OUTLIERS IDENTIFIED BY ASTERISKS

J	DELTA	H/R	G	J	DELTA	H/R	G
1	1.4132	91.12	0.70	34	1.2653	89.42	0.85
2	1.5222	96.05	0.75	35	0.9751	90.48	0.90
3	1.3226	96.32	0.80	36	2.8389	85.05	0.95
4	3.1165	102.30	0.85	37	0.5015	96.03	1.00
5	0.7199	92.57	0.90	38	0.6559	98.22	1.05
6	0.6203	95.64	0.95	39	0.2377	95.63	1.10
7	0.6784	91.44	1.00	40	3.4797	106.36	1.15
8	0.4002	93.12	1.05	41	0.0979	96.23	1.20
9	0.2635	94.08	1.10	42	0.9158	90.71	1.25
10	0.5249	91.90	1.15	43	0.2202	93.94	1.30
11	0.1091	96.60	1.20	44	0.3527	93.46	1.35
12	0.1499	98.14	1.25	45	0.4353	93.43	1.40
13	0.6519	101.35	1.30	46	0.1363	95.65	1.45
14	0.1888	99.55	1.35	47	0.1967	96.50	1.55
15	0.0746	95.74	1.40	48	0.4257	101.94	1.60
16	0.0263	98.32	1.45	49	2.4135	107.98	1.65
17	0.1421	99.93	1.50	50	0.3134	99.45	1.70
18	0.7313	103.21	1.55	51	4.1166	111.35	1.75
19	0.6055	102.83	1.60	52	0.9969	104.31	1.80
20	0.3148	99.79	1.70	53	1.1297	103.63	1.95
21	1.1413	104.90	1.80	54	1.8467	101.49	2.15
22	0.7922	98.58	1.85	55	3.9882	111.52	2.20
23	0.9621	103.24	1.90	56	8.0630	117.56	2.55
24	1.1990	104.12	1.95	57	4.9986	110.36	2.60
25	1.6294	101.03	2.10	58	20.3487 *	98.34	3.55
26	6.1920	90.37	2.15	59	1.3188	96.30	0.80
27	2.0352	103.30	2.20	60	4.3616	82.15	0.85
28	2.6309	87.15	0.50	61	4.0832	82.72	0.90
29	11.1150 *	109.02	0.55	62	3.4158	83.98	0.95
30	2.3074	96.89	0.65	63	0.7541	91.02	1.00
31	1.3906	91.58	0.70	64	0.3372	94.05	1.05
32	2.7500	85.14	0.75	65	2.0682	90.08	1.55
33	1.9632	87.12	0.80	66	2.9773	102.14	2.35

* OUTLIER IDENTIFIED AT 0.01 SIGNIFICANCE LEVEL
 SAMPLE SIZE IS 66
 NO OF OUTLIERS IS 2
 DELSTAR = 8.7115

DATA BEFORE IDENTIFICATION OF OUTLIERS			DATA AFTER DELETION OF OUTLIERS		
VARIABLES	MEAN	S.D.	MEAN	S.D.	
H/R	97.1806	7.2956	96.9775	7.2544	
G	1.3773	0.5767	1.3562	0.5069	

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